Assignment 3

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Faculty: FES Unit Code: MEME15203

Course: MAC Unit Title: Statistical Inference Year: 1,2 Lecturer: Dr Yong Chin Khian

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Due by:

Q1. Consider a random sample of size n from a single-parameter pareto distribution, $X_i \sim SP(\alpha = 2, \theta)$. Find the UMVUE of the p^{th} percentile.

(10 marks)

Ans.

$$f(x_i) = \frac{2\theta^2}{x_i^3}, x_i > \theta$$

$$f(x, x_2, \dots, x_n) = 2^n \theta^{2n} \prod_{i=1}^n x_i^{-3}, x_{(1)} > \theta$$

$$= g(s, \theta) h(x_1, \dots, x_n)$$

where $g(s,\theta) = \theta^{2n} I(X_{(1)}) > \theta$ and $h(x_1, \dots, x_n) = 2^n \prod_{i=1}^n x_i^{-3}$, By factorization Theorem, $S = X_{(1)}$ is a sufficient statistic for θ .

$$f_S(s) = nf_X(s)[1 - F_X(s)]^{(n-1)} = \frac{2n\theta^2}{s^3} \left(\frac{\theta}{s}\right)^{2(n-1)} = \frac{2n\theta^{2n}}{s^{2n+1}}, s > \theta$$

$$\Rightarrow S \sim SP(\alpha = 2n, \theta)$$

$$E[u(S)] = \int_{\theta}^{\infty} u(s) \frac{2n\theta^{2n}}{s^{2n+1}} ds = 0 \ \forall \theta$$

$$\Rightarrow \int_{\theta}^{\infty} u(s) s^{-2n} ds = 0 \ \forall \theta$$

$$\frac{d}{d\theta} \int_{\theta}^{\infty} u(s) s^{-2n} ds = u(\theta) \theta^{-2n} = 0 \ \forall \theta$$
This implies $u(\theta) = 0$ for all θ , so $S = X_{(1)}$ is complete.

$$E(S) = \frac{2n\theta}{2n-1}$$

$$P(X > \pi_p) = 1 - p$$
$$\left(\frac{\theta}{\pi_p}\right)^2 = 1 - p$$
$$\pi_p = \frac{\theta}{\sqrt{1-p}}$$

Let
$$S_1 = \frac{(2n-1)S}{2n\sqrt{1-p}}$$
, then $E(S_1) = \frac{(2n-1)E(S)}{2n\sqrt{1-p}} = \frac{\theta}{\sqrt{1-p}}$

Let $S_1 = \frac{(2n-1)S}{2n\sqrt{1-p}}$, then $E(S_1) = \frac{(2n-1)E(S)}{2n\sqrt{1-p}} = \frac{\theta}{\sqrt{1-p}}$ Since $S_1 = \frac{(2n-1)S}{2n\sqrt{1-p}}$ is a function of css for θ and unbisaed for π_p . Then, $S_1 = \frac{(2n-1)S}{2n\sqrt{1-p}}$ is the UMVUE of the p^{th} percentile.

- Q2. Suppose that $X_1, ..., X_n$ is a random sample from a Negative Binomial distribution, $X_i \sim \text{NB}(\mathbf{r} = 6, \theta)$,
 - (a) Show that the p.d.f. of X belongs to the regular exponential family.
 - (b) Based on the answer in 1, find a complete and sufficient statistic for θ .
 - (c) Find the UMVUE of $\left[\frac{4\theta}{1-4(1-\theta)}\right]^{6n}$.

(15 marks)

Ans.

- (a) $f(x) = \binom{x-1}{r-1}\theta^6(1-\theta)^{x-6} = \binom{x-1}{r-1}\left(\frac{\theta}{1-\theta}\right)^6e^{x\ln(1-\theta)} = c(\theta)h(x)e^{q(\theta)t(x)}$ where $c(\theta) = \left(\frac{\theta}{1-\theta}\right)^6$, $h(x) = \binom{x-1}{r-1}$, $q(\theta) = \ln(1-\theta)$, and t(x) = x. Thus the p.d.f. of X belongs to the regular exponential family.
- (b) Since the p.d.f. of X belongs to the regular exponential family, thus by the theorem, $S = \sum_{i=1}^{n} X_i$ is a c.s.s of θ
- (c) Let $E(t^S) = \left[\frac{4\theta}{1-4(1-\theta)}\right]^{6n}$ As $S \sim NB(6n, \theta)$, $E(t^S) = E(e^{S \ln t}) = \left(\frac{t\theta}{1-t(1-\theta)}\right)^{6n}$ $\Rightarrow t = 4$ Since 4^S is a function of the c.s.s. of θ which is an UE of $\left(\frac{4\theta}{1-4(1-\theta)}\right)^{6n}$, thus 4^S is the UMVUE of $\left(\frac{4\theta}{1-4(1-\theta)}\right)^{6n}$.
- Q3. Consider a random sample of size n from a gamma distribution $X_i \sim GAM(\alpha, \theta)$ and let $\bar{X} = (1/n) \sum X_i$ and $\tilde{X} = (\prod X_i)^{1/n}$ be the sample mean and geometric mean respectively.
 - (a) Show that \bar{X} and \tilde{X} are jointly complete and sufficient for θ and α .
 - (b) Find the UMVUE of $\mu = \alpha \theta$.
 - (c) Find the UMVUE of μ^n ..
 - (d) Show that \bar{X} and T are stochastically independent random variables.

(20 marks)

Ans.

(a)
$$f(x_i) = \frac{1}{\Gamma(\alpha)\theta^{\alpha}} x_i^{\alpha-1} e^{-x_i/\theta}$$
$$= \frac{1}{\Gamma(\alpha)\theta^{\alpha}} x_i^{\alpha} e^{\alpha \ln x_i} e^{-x_i/\theta}$$

$$=\frac{1}{\Gamma(\alpha)\theta^{\alpha}}x_i^{-1}e^{\alpha\ln x_i-x_i/\theta}$$

Thus $X_i \sim GAM(\alpha, \theta)$ belongs to REC family where $c(\theta) = \frac{1}{\Gamma(\alpha)\theta^{\alpha}}, h(x) = x_i^{-1}, t_1(X_i) = \ln X_1, t_2(X_i) = X_1, q_1(\theta) = \alpha$ and $q_2(\theta) = \theta$.

Thus, $\sum \ln X_i$ and $\sum X_i$ are jointly complete and sufficient of α and θ . Since \bar{X} is a function of $\sum X_i$ and \tilde{X} is a function of $\sum \ln X_i$, thus \bar{X} and \tilde{X} are jointly complete and sufficient for θ and α .

- (b) $\sum X_i \sim GAM(n\alpha, \theta)$ $2 \sum X_i/\theta \sim \chi^2(2n\alpha)$ $E(2n\bar{X}/\theta) = 2n\alpha$ $E(\bar{X}) = \alpha\theta$ Since \bar{X} is a CSS of θ and α , thus, \bar{X} is the UMVUE of $\mu = \alpha\theta$.
- (c) $E(\prod X_i) = (EX_i)^n = \mu^n$ Thus, $\prod X_i$ is an UE of μ . Since $\prod X_i = \tilde{X}^n$, then $\prod X_i$ is the UMVUE of μ^n .
- (d) $T = u(X_1, X_2, ..., X_n)$ $= u(cX_1, cX_2, ..., cX_n)$ $= \frac{(cX_1 + cX_2 + ..., + cX_n)/n}{(cX_1 \cdot cX_2 \cdot ... cX_n)^{1/n}}$ $= \frac{c(X_1 + X_2 + ..., + X_n)/n}{c(X_1 \cdot X_2 \cdot ... X_n)^{1/n}}$ $= \frac{\bar{X}}{\bar{X}}$

Since \bar{X} is a CSS of θ and the distribution of T does not depend on θ . thus by BASU theorem, \bar{X} and T are stochastically independent random variables.

Q4. %%%% Suppose that X_1, \ldots, X_n is a random sample from a Poisson distribution, $X_i \sim \text{POI}(\lambda)$. Find the UMVUE of $P(X_1 + X_2 = 0 \text{ or } 1) = (1 + 2\lambda)e^{-2\lambda}$ using Rao-Blackwell theorem.

(10 marks)

Ans.

 $f(x) = \frac{1}{lambda} e^{-x/\lambda} = c(\lambda)h(x)e^{t(x)q(\lambda)}$ which is in a member of REC. Hence $S = \sum X_i$ is a CSS of λ . Let

$$T = \begin{cases} 1, & X_1 + X_2 = 0 \text{ or } 1\\ 0, & \text{otherwise} \end{cases}.$$

 $E(T)=P(X_1+X_2=0 \text{ or } 1)=e^{-2\lambda}+2\lambda e^{-2\lambda}=(1+2\lambda)e^{-2\lambda}.$ Thus T is and unbiased estimator of $(1+2\lambda)e^{-2\lambda}$. Since S is CSS of λ . Hence by Rao-Blackwell theorem, $T^*=E(T|S)$ is an UMVUE of $(1+2\lambda)e^{-2\lambda}$.

Rao-Blackwell theorem,
$$T^* = E(T|S)$$
 is an UMVUE of $(1+2\lambda)e^{-2\lambda}$.
$$E\left[T|\sum_{i=0}^{n}X_{i}=s\right]$$

$$= 1 \cdot P[X_{1} + X_{2} = 0 \text{ or } 1|X_{1} + X_{2} + \cdots + X_{n} = s]$$

$$= \frac{P(X_{1} + X_{2} = 0, X_{3} + \cdots + X_{n} = s)}{P(X_{1} + \cdots + X_{n} = s)} + \frac{P(X_{1} + X_{2} = 1, X_{3} + \cdots + X_{n} = s - 1)}{P(X_{1} + \cdots + X_{n} = s)}$$

$$= \frac{P(X_{1} + X_{2} = 0) \times P(X_{3} + \cdots + X_{n} = s)}{P(X_{1} + \cdots + X_{n} = s)} + \frac{P(X_{1} + X_{2} = 1) \times P(X_{3} + \cdots + X_{n} = s - 1)}{P(X_{1} + \cdots + X_{n} = s)} \text{Since} \quad X_{1}, X_{2}, \dots, X_{n} \quad \text{are} \quad \text{independent.}$$

$$= \frac{e^{-2\lambda}[(n-2)\lambda]^{s}e^{-(n-2)\lambda}}{(n\lambda)^{s}s!e^{-n\lambda}/s!} + \frac{2\lambda e^{-2\lambda}[(n-2)\lambda]^{s-1}e^{-(n-2)\lambda}}{(n\lambda)^{s}(s-1)!e^{-n\lambda}/s!}$$

$$= (\frac{n-2}{n})^{s} + (\frac{n-2}{n})^{s}(\frac{2s}{n-2})$$

Q5. Consider a random sample of size n from a distribution with pdf

$$f(x;\theta) = \frac{(\ln \theta)^x}{\theta x!}, x = 1, 1, \dots; \theta > 1, \text{ zero, otherwise.}$$

- (a) Find a complete sufficient statistic for θ .
- (b) Find the UMVUE of $\ln \theta$.
- (c) Find the UMVUE of $(\ln \theta)^2$.

(15 marks)

Ans.

- (a)Let $\mu = \ln \theta$ $f(x;\theta) = \frac{\mu^x}{e^{\mu}x!} = e^{-\mu} \frac{1}{x!} e^{x \ln(\mu)} = c(\mu) h(x) e^{q(\mu)t(x)}$ where $c(\mu) = e^{-\mu}$, $h(x) = \frac{1}{x!}$, $q(\mu) = \ln(\mu)$; t(x) = x. Thus, $f(x;\theta)$ belong to $REC(\theta)$ and hence $S = \sum X_i$ is a complete sufficient statistic for $\mu = \ln \theta$.
- (b) $X \sim POI(\mu = \ln \theta)$, $E(X) = \ln \theta$ and $E\left(\frac{S}{n}\right) = E(\bar{X}) = \ln \theta$. Since \bar{X} is a function of CSS and unbiased for $\ln \theta$, therefore, by Lehmann Scheffe theorem, \bar{X} is the UMVUE of $\ln \theta$.
- (c) $E\left[\bar{X}^2 \frac{\bar{X}}{n}\right] = \frac{\mu}{n} + \mu^2 \frac{\mu}{n} = \mu^2$. Since $\bar{X}^2 \frac{\bar{X}}{n}$ is a function of CSS and unbiased for $\mu^2 = (\ln \theta)^2$, therefore, by Lehmann Scheffe theorem, $\bar{X}^2 \frac{\bar{X}}{n}$ is the UMVUE of $(\ln \theta)^2$.

Q6. Show that $X \sim N(0, \theta)$ is not a complete family.

(10 marks)

Ans.

$$X \sim N(0,\theta)$$

Let U(X) = X, Then E[U(S)] = E(X) = 0, but $X \neq 0$. Thus, $X \sim N(0, \theta)$ is not a complete family.

Q7. Let $X_1, X_2, ..., X_n$ be random sample of size n from a Gamma distribution with probability density function

$$\frac{1}{\theta^2}xe^{-x/\theta}, x > 0$$

zero otherwise. Find the UMVUE of $\gamma = P(X > t)$ using Rao-Blackwell theorem. (20 marks)

Ans.

Let

$$T = \begin{cases} 1, & X_1 > t \\ 0, & \text{otherwise} \end{cases}.$$

Then, E(T) = P(X > t). Thus T is and unbiased estimator of γ .

 $f(x) = \frac{1}{\theta^2} e^{-x/\theta} = c(\theta) h(x) e^{q(\theta)t(x)}$, where $c(\theta) = \frac{1}{\theta^2}$, h(x) = 1, $q(\theta) = \frac{1}{\theta}$, and t(x) = x, hence f(x) is a member of $REC(\theta)$ and $S = \sum_{i=1}^n X_i$ is a complete sufficient statistics for θ .

Thus, by Rao-Blackwell therem, $T^* = E(T|S)$ is an UMVUE of γ .

$$f_{X_1,S}(x_1,s) = f_{X_1,S_1}(x_1,s-x_1) \text{ where } S_1 = X_2 + \cdots, X_n \sim gamma(2n-2,\theta)$$

$$= f_{X_1}(x_1)f_{S_1}(s-x_1)$$

$$= \frac{1}{\theta^2}x_1e^{-x_1/\theta}\frac{1}{\Gamma(2n-2)\theta^{2n-2}}(s-x_1)^{2n-3}e^{-(s-x_1)/\theta}$$

$$= \frac{1}{\Gamma(2n-2)\theta^{2n}}x_1(s-x_1)^{2n-3}e^{-s/\theta}$$

$$f_{X_1|S}(x_1) = kx_1(s - x_1)^{2n-3}, 0 < x_1 < s$$
Let $z = s - x_1, dz = -dx$

$$\int_s^0 k(s - z)z^{2n-3}dx_1 = 1$$

$$\int_s^0 k(sz^{2n-3} - z^{2n-2})dz = 1$$

$$k \left[\frac{sz^{2n-2}}{2n-2} - \frac{z^{2n-1}}{2n-1} \right]_0^s = 1$$

$$k \left[\frac{s^{2n-1}}{2n-2} - \frac{s^{2n-1}}{2n-1} \right] = 1$$

$$k \left[\frac{(2n-1)s^{2n-1} - (2n-2)s^{2n-1}}{(2n-1)(2n-2)} \right] = 1$$

$$k \left[\frac{(2n-1)s^{2n-1} - (2n-2)s^{2n-1}}{(2n-1)(2n-2)} \right] = 1$$

$$k = (2n-2)(2n-1)s^{1-2n}$$

$$\therefore f_{X_1|S}(x_1) = (2n-2)(2n-1)s^{1-2n}x_1(s-x_1)^{2n-3}, 0 < x_1 < s$$

$$E[T|S] = P[X_1 > t|s]$$

$$= \int_t^s (2n-2)(2n-1)s^{1-2n}x_1(s-x_1)^{2n-3}dx_1$$
Let $z = s - x_1$, $dz = -dx$

$$= (2n-2)(2n-1)s^{1-2n} \int_{s-t} t^0(s-z)z^{2n-3}(-dz)$$

$$= (2n-2)(2n-1)s^{1-2n} \left[\frac{sz^{2n-2}}{2n-2} - \frac{z^{2n-1}}{2n-1} \right]_0^{s-t}$$

$$= (2n-2)(2n-1)s^{1-2n} \left[\frac{s(s-t)^{2n-2}}{2n-2} - \frac{(s-t)^{2n-1}}{2n-1} \right]$$

$$= (2n-2)(2n-1) \left[\frac{\left(\frac{s-t}{s}\right)^{2n-2} - \left(\frac{s-t}{s}\right)^{2n-1}}{2n-1} \right]$$

$$= (2n-2)(2n-1) \left[\frac{(2n-1)\left(\frac{s-t}{s}\right)^{2n-2} - (2n-2)\left(\frac{s-t}{s}\right)^{2n-1}}{(2n-2)(2n-1)} \right]$$

$$= (2n-1)\left(\frac{s-t}{s}\right)^{2n-2} - (2n-2)\left(\frac{s-t}{s}\right)^{2n-1}$$